

Numerical experiment with the construction of a control operator applied in ECGM algorithm

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Abstract: In this paper, we constructed a control operator, G, which enables an Extended Conjugate Gradient Method (ECGM) to be employed in solving for the optimal control and trajectories of continuous time linear regulator problems. Similar operators constructed in the past by various authors have limited application. This call for the construction of the control operator that is aimed at taking care of any of the Mayer’s, Lagrange’s and Bolza’s cost form of linear regulator problems. The authors of this paper desire that, with the construction of the operator, one will circumvent the difficulties undergone using the classical methods and its application will further improve the result of the Extended Conjugate Gradient Method in solving this class of optimal control problem. The constructed Linear Control Operator is applied in ECGM algorithm to solve Continuous-Time Linear Regulator Problems with the convergence profile showing the efficiency of the operator.

Keywords: Continuous Linear Regulator Problem, Control Operator, Extended Conjugate Gradient Method, Optimal Control

1. Introduction

Consider the linear regulator problem of the Bolza type in [1] and [6] as:

Problem (1):

$$J(x, t_f, u(\cdot)) = \frac{1}{2}x^T(t_f)Hx(t_f) + \frac{1}{2}\int_0^{t_f}\{x^T(t)Q(t)x(t) + u^T(t)R(t)u(t)\}dt \quad (1.1)$$

subject to the differential state equation

$$\dot{x}(t) = A(t)x(t) + B(t)u(t), \quad 0 \leq t \leq t_f, \quad (1.2)$$

where H and Q(t) are real symmetric positive semi-definite $n \times n$ matrices. R(t) is a real symmetric positive definite $m \times m$ matrix, the initial time, and the final time, t_f , are specified. $x(t)$ is an n -dimensional state vector, $u(t)$ is the m -dimensional plant control input vector. $u(t)$ and $x(t)$ are not constrained by any boundaries. A and B are specified constants which are not necessarily positive. For $H = 0$, (1.1) is called a Lagrange problem, but if $Q(t) = R(t) = 0$, it is called a Mayer problem.

The form (1.1) may be rewritten as:

$$J = \int_0^{t_f} \left\{ \frac{d}{dt} \left(\frac{1}{2}x^T(t_f)Hx(t_f) \right) \right\} dt + \frac{1}{2}\int_0^{t_f}\{x^T(t)Q(t)x(t) + u^T(t)R(t)u(t)\}dt \quad (1.3)$$

$$J = \int_0^{t_f}\{x^T(t)H\dot{x}(t)\} dt + \frac{1}{2}\int_0^{t_f}\{x^T(t)Q(t)x(t) + u^T(t)R(t)u(t)\}dt \quad (1.4)$$

$$J = \int_0^{t_f} \left\{ x^T(t)H\dot{x}(t) + \frac{1}{2}x^T(t)Q(t)x(t) + \frac{1}{2}u^T(t)R(t)u(t) \right\} dt.$$

As customary with penalty function techniques, constrained problem equations (1.2) and (1.4) may be put into the following equivalent form:

$$\langle z, Gz \rangle_k = \min_{(x,u)} \int_0^{t_f} \{x^T(t)H\dot{x}(t) + \frac{1}{2}x^T(t)Q(t)x(t) + \frac{1}{2}u^T(t)R(t)u(t) + \mu \| \dot{x}(t) - A(t)x(t) - B(t)u(t) \|^2\} dt \quad (1.5)$$

where $\mu > 0$ is the penalty parameter and $\|\dot{x}(t) - A(t)x(t) - B(t)u(t)\|^2$ is the penalty term.

Let us denote by \tilde{k} the product space

$$\tilde{k} = \mathcal{H}[0, t_f] \times \ell_2[0, t_f] \tag{1.6}$$

is the product space of Sobolev space $\mathcal{H}[0, t_f]$ of absolutely continuous function $x(\cdot)$ such that, both $x(\cdot)$ and $\dot{x}(\cdot)$ are square integrable over the finite interval $[0, t_f]$ and the Hilbert space $\ell_2[\alpha, \beta]$ of equivalence classes of real valued functions on $[\alpha, \beta]$ with norm defined by:

$$\|f(\cdot)\|_{\ell_2[\alpha, \beta]} = \left(\int_{\alpha}^{\beta} |f(t)|^2 dt \right)^{\frac{1}{2}}, f(\cdot) \in \ell_2[\alpha, \beta] \tag{1.7}$$

Then, the inner product $\langle \cdot, \cdot \rangle_{\tilde{k}}$ on \tilde{k} is given by

$$\langle \cdot, \cdot \rangle_{\tilde{k}} = \langle \cdot, \cdot \rangle_{\mathcal{H}[0, t_f]} + \langle \cdot, \cdot \rangle_{\ell_2[0, t_f]} \tag{1.8}$$

Suppose $z(t) \in \tilde{k}$ denotes the ordered pair

$$z^T(t) = (x(t), u(t)); x(t) \in \mathcal{H}[0, t_f], u(t) \in \ell_2[0, t_f] \tag{1.9}$$

then, we seek to determine the operator G on \tilde{k} such that

$$\langle z, Gz \rangle_{\tilde{k}} = \int_0^{t_f} \{x^T(t)H\dot{x}(t) + \frac{1}{2}x^T(t)Q(t)x(t) + \frac{1}{2}u^T(t)R(t)u(t)\} dt \tag{1.10}$$

$$+ \mu \|\dot{x}(t) - A(t)x(t) - B(t)u(t)\|^2 dt$$

where \tilde{k} is suitably chosen Hilbert space. The control operator, G, is then utilized in the iterative framework of the CGM in order to arrive at a solution of problem (P1). We provide a recapitulation of the formal CGM in the next section for the sake of completeness.

2. Conjugate Gradient Method Algorithm

The Conjugate Gradient Method (CGM) algorithm for iteratively locating the minimum x^* of $f(x)$ in \mathcal{H} as described by [4] is as follows:

Step 1: Guess the first element $x_0 \in \mathcal{H}$ and compute the remaining members of the sequence with the aid of the formulae in the steps 2 through 6.

$$\text{Step 2: Compute the descent direction } p_0 = -g_0 \tag{2.1a}$$

$$\text{Step 3: Set } x_{i+1} = x_i + \alpha_i p_i; \text{ where } \alpha_i = \frac{\langle g_i, g_i \rangle_{\mathcal{H}}}{\langle p_i, Gp_i \rangle_{\mathcal{H}}} \tag{2.1b}$$

$$\text{Step 4: Compute } g_{i+1} = g_i + \alpha_i Gp_i \tag{2.1c}$$

$$\text{Step 5: Set } p_{i+1} = -g_{i+1} + \beta_i p_i; \beta_i = \frac{\langle g_{i+1}, g_{i+1} \rangle_{\mathcal{H}}}{\langle g_i, g_i \rangle_{\mathcal{H}}} \tag{2.1d}$$

Step 6: If $g_i = 0$ for some i , then, terminate the sequence; else set $i = i + 1$ and go to step 3.

In the iterative steps 2 through 6 above, p_i denotes the descent direction at i -th step of the algorithm, α_i , is the step

length of the descent sequence $\{x_i\}$ and g_i denotes the gradient of F at x_i . Steps 3, 4 and 5 of the algorithm reveal the crucial role of the linear operator G in determining the step length of the descent sequence and also in generating a conjugate direction of search. Applicability of the algorithm thus depends solely on the explicit knowledge of the operator G. Generally, for discrete optimization problems, G is readily determined (see [4, pp. 51-53]); and such problem enjoys the beauty of the CGM as a computational scheme since the CGM exhibits quadratic convergence and requires only a little more computation per iteration. [2] opined that these properties make the CGM a fascinating computational technique with a strong appeal for implementation on the digital computer.

However, for the type of constrained continuous linear time regulator problem (P1) discussed in this paper, application of the CGM algorithm as presented is hindered because then, the equivalent of operator G which satisfies (P1) in this sense of (1.13) is not readily found and construction of such operator is the main aim of this paper. The construction of such an operator is not new. For instance in [5], the authors constructed the control operator for the following related problem respectively as:

Problem (2)

$$\text{Min}_{(x,u)} \int_{t_0}^{t_f} \{ax^2(t) + bu^2(t)\} dt \tag{2.2}$$

Subject to

$$\dot{x}(t) = cx(t) + du(t), \quad t_0 \leq t \leq t_f, \tag{2.3}$$

$$x(t_0) = x_0; \tag{2.4}$$

where a, b, c and d specified constants such that $a > 0, b > 0; x_0, t_f$ are given, $\dot{x}(t)$ denotes the derivative of the state $x(\cdot)$ with respect to time, and $u(\cdot)$ is the control vector. The resulting control operator G is of the form:

$$(G_{11}x)(t) = -\mu[\dot{x}(0) - Cx(0)]\text{Sinh}(t) + \mu \int_0^t [\dot{x}(s) - Cx(s)]\text{Cosh}(t-s) ds$$

$$- \int_0^t [(A + \mu C^2)x(s) - \mu C\dot{x}(s)]\text{Sinh}(t-s) ds$$

$$+ [(a + \mu C^2)x(0) - \mu C\dot{x}(0)]\text{Cosh}(t)$$

$$+ \frac{\text{Sinh}(t)}{\text{Sinh}(T)} \{ (a + \mu C^2)x(T) - \mu C\dot{x}(T) + \mu \text{Sinh}(T)[\dot{x}(0) - Cx(0)] \} \tag{2.5}$$

$$- \mu \int_0^T [\dot{x}(s) - Cx(s)]\text{Cosh}(T-s) ds$$

$$+ \int_0^T [(A + \mu C^2)x(s) - \mu C\dot{x}(s)]\text{Sinh}(T-s) ds$$

$$- [(A + \mu C^2)x(0) - \mu C\dot{x}(0)]\text{Cosh}(T), \quad 0 \leq t \leq T$$

$$(G_{21}x)(t) = \mu CDx(t) - \mu D\dot{x}(t), \quad 0 \leq t \leq T \tag{2.6}$$

$$\begin{aligned}
 (G_{12}u)(t) &= \mu Du(0) \text{Sinh}(t) - \mu \int_0^t Du(s) \text{Cosh}(t-s) ds \\
 &+ \mu \int_0^t CDu(s) \text{Sinh}(t-s) ds + \mu CDu(0) \text{Cosh}(t) \\
 &+ \frac{\text{Sinh}(t)}{\text{Sinh}(T)} \{ \mu CDu(T) - \mu Du(0) \text{Sinh}(T) \} \\
 &+ \mu \int_0^T Du(s) \text{Cosh}(T-s) ds + \mu \int_0^T CDu(s) \text{Sinh}(T-s) ds \\
 &+ \mu CDu(0) \text{Cosh}(T) \}, 0 \leq t \leq T
 \end{aligned} \tag{2.7}$$

$$(G_{22}u)(t) = Bu(t) + \mu D^2u(t), 0 \leq t \leq T \tag{2.8}$$

Based on 2.5 – 2.8 above, one can obtain the desired Gp_i to apply 2.1c which would henceforth allow us to exploit the simplicity of the conjugate gradient method algorithm.

Similarly, Aderibigbe, (1993), constructed a control operator for a linear regulator problem with delay parameter as:

Problem (3)

$$\text{Min}_{(x,u)} \int_{t_0}^{t_f} \{ Ax^2(t) + Bu^2(t) \} dt \tag{2.9}$$

Subject to

$$\dot{x}(t) = C_1x(t) + C_2x(t-r) + Du(t), \quad t_0 \leq t \leq t_f, \tag{2.10}$$

$$x(t) = h(t); \quad -r \leq t \leq 0 \tag{2.11}$$

where $A, B > 0$; the delay parameter $r > 0$ and t_f are given; C_1, C_2 and D are specified constants which are not necessarily positive and $h(t)$ is a given piecewise continuous function which is of exponential order on $[-r, 0]$. For detail, see [2].

3. Main Result

Our results for problem (P1) are contained in the following theorem:

3.1. Theorem 1

Let us denote by \tilde{k} the product space

$$\tilde{k} = W^{1,2}[t_0, t_f] \times \ell^2[t_0, t_f] \tag{3.1}$$

of the Sobolev space of absolutely continuous function $x(\cdot)$ on $[t_0, t_f]$ and the usual Hilbert space $\ell^2[t_0, t_f]$ of Lebesgue measurable, real-valued functions which are square integrable on the closed interval $[t_0, t_f]$. Then, for $\mu > 0$ there exists a control operator, G , with $G: \tilde{k} \times \tilde{k} \rightarrow \mathcal{R}$ such that

$$\langle z(t), Gz(t) \rangle_{\tilde{k}} = \int_{t_0}^{t_f} \{ x^T(t) H \dot{x}(t) + \frac{1}{2} x^T(t) Q(t) x(t) + \frac{1}{2} u^T(t) R(t) u(t) + \mu \| Ax(t) + Bu(t) - \dot{x}(t) \|^2 \} dt \tag{3.2}$$

where $z(t) = (x(t), u(t))^T$. Furthermore, G is explicitly given by

$$(Gz(t)) \equiv \begin{pmatrix} G_{11} & G_{12} \\ G_{21} & G_{22} \end{pmatrix} = \begin{pmatrix} (G_{11}x)(t) + (G_{12}u)(t) \\ (G_{21}x)(t) + (G_{22}u)(t) \end{pmatrix} \tag{3.3}$$

where the composite operators $G_{ij}, 1 \leq i, j \leq 2$ are given as follows:

$$\begin{aligned}
 (G_{11}x_2)(t) &= -\mu \dot{x}_2(0) \text{Sinh}(t_f) + \int_0^{t_f} \mu \dot{x}_2(s) \text{Cosh}(t_f - s) ds - \int_0^{t_f} \text{Sinh}(t_f - s) \\
 &\times \left[\left(\frac{1}{2} Q + \mu A^2 \right) x(s) + (H - 2\mu A) \dot{x}_2(s) \right] ds + \\
 &\left[\left(\frac{1}{2} Q + \mu A^2 \right) x_2(0) + (H - 2\mu A) \dot{x}_2(0) \right] \\
 &\times \text{Cosh}t + \frac{\text{Sinh}t}{\text{Sinh}t_f} \{ \left(\frac{1}{2} Q + \mu A^2 \right) x_2(t_f) + (H - 2\mu A) \dot{x}_2(t_f) - \left[\left(\frac{1}{2} Q + \mu A^2 \right) x_2(0) \right. \\
 &\left. + (H - 2\mu A) \dot{x}_2(0) \right] \text{Cosh}t_f + \mu \dot{x}_2(0) \text{Sinh}t_f - \int_0^{t_f} \mu \dot{x}_2(s) \text{Cosh}(t_f - s) ds \\
 &\left. + \int_0^{t_f} \text{Sinh}(t_f - s) \left[\left(\frac{1}{2} Q + \mu A^2 \right) x(s) + (H - 2\mu A) \dot{x}_2(s) \right] ds \}; 0 \geq t \geq t_f
 \end{aligned} \tag{3.4}$$

$$\begin{aligned}
 (G_{12}u_2)(t) &= \\
 2\mu ABu_2(0) \text{Cosh}t - \int_0^{t_f} 2\mu ABu_2(s) \text{Sinh}(t_f - s) ds + \\
 \frac{\text{Sinh}t}{\text{Sinh}t_f} \{ 2\mu ABu_2(t_f) \\
 - 2\mu ABu_2(0) \text{Cosh}(t_f) + \int_0^{t_f} 2\mu ABu_2(s) \text{Sinh}(t_f - s) ds \}; 0 \geq t \geq t_f
 \end{aligned} \tag{3.5}$$

$$(G_{21}x_2)(t) = -2\mu B \dot{x}_2(t); 0 \geq t \geq t_f \tag{3.6}$$

$$(G_{22}u_2)(t) = \left(\frac{1}{2} R + \mu B^2 \right) u_2(t); 0 \geq t \geq t_f \tag{3.7}$$

Proof of Theorem 3.1

In an attempt to proof the above theorem, we need the following fundamentals:

3.2. Lemma 1

Let $n \geq 0$ be an integer and suppose $C^n(0, T)$ denotes the space of all real-valued functions $y(t), 0 \leq t \leq T$ which are n -times continuously differentiable on $[0, T] \subset \mathcal{R}$ with the norm $\|y\|_n$ given by

$$\|y\|_n = \sum_{i=0}^n \max_{0 \leq t \leq T} |y^{(i)}(t)|, \tag{3.8}$$

where $|y^{(i)}(t)|$ denotes the i th derivative of $y(t)$. Then if $\alpha(t)$ and $\beta(t)$ are continuous in $[a, b]$ and if $\int_a^b [\alpha(t)y(t) + \beta(t)\dot{y}(t)] dt = 0$ for every function $g(t) \in C^1(a, b)$ such that $g(a) = g(b) = 0$, then $\beta(t)$ is

differentiable and $\frac{d}{dt}(\beta(t)) = \alpha(t)$ for all $t \in [a, b]$.

Proof

For our subsequent development we shall associate with the right hand side of (1.10) the functional $R_\mu(z_1, z_2)$ defined by

$$R_\mu(z_1, z_2) = \int_0^{t_f} \left\{ \frac{1}{2} x_1(t) Q x_2(t) + \frac{1}{2} u_1(t) R u_2(t) + x_1(t) H \dot{x}_2(t) + \mu \dot{x}_1(t) \dot{x}_2(t) - \mu A x_1(t) \dot{x}_2(t) - \mu B u_1(t) \dot{x}_2(t) - \mu A x_1(t) \dot{x}_2(t) + \mu A^2 x_1(t) x_2(t) - \mu B u_1(t) \dot{x}_2(t) \right\} dt$$

$$+ \mu A B x_1(t) u_2(t) - \mu B u_1(t) \dot{x}_2(t) + \mu A B x_1(t) u_2(t) + \mu B^2 u_1(t) u_2(t) \} dt \tag{3.9}$$

where $z_1^T = (x_1(t), u_1(t))$, $z_2^T = (x_2(t), u_2(t))$ are the ordered triple pair which belong to the space \tilde{k} defined by (3.1). It follows that; the form (3.9) is equivalent to (3.2) under the equivalent relationships:

$$\left. \begin{aligned} x_1(t) &\equiv x_2(t) = x(t), & 0 \geq t \geq t_f, \\ \dot{x}_1(t) &\equiv \dot{x}_2(t) = \dot{x}(t), & 0 \geq t \geq t_f, \\ u_1(t) &\equiv u_2(t) = u(t), & 0 \geq t \geq t_f. \end{aligned} \right\} \tag{3.10}$$

For proof, see [7]. We then have the following proposition:

3.3. Proposition 1

$R_\mu(z_1, z_2)$ is a bounded, bilinear, self-adjoint form on \tilde{k} .

Proof:

Bilinearity and self-adjointness of $R_\mu(z_1, z_2)$ is clear from its definition; and its boundedness follows from the fact that $z_i(t) = (x_i(t), u_i(t))^T, i = 1, 2$ is bounded.

3.4. Remarks

By virtue of proposition above and a consequence of the Reiss representation theorem on Hilbert spaces [8], it follows that $R_\mu(z_1, z_2)$ induces a uniquely determined, bounded linear operator G say on \tilde{k} with the representation

$$R_\mu(z_1, z_2) = \langle G z_1, z_2 \rangle_{\tilde{k}} = \langle z_1, G z_2 \rangle_{\tilde{k}} = R_\mu(z_1, z_2) \tag{3.11}$$

where it is clear that G is also self-adjoint on \tilde{k} since $R_\mu(z_1, z_2)$ is.

Let us now consider the equivalence

$$\langle z_1, G z_2 \rangle = R_\mu(z_1, z_2) \tag{3.12}$$

This is convenient for our subsequent developments.

Then, let $z_2(t) = (x_2(t), u_2(t))^T$, then we can write

$$(G z_2(t)) \equiv \begin{pmatrix} G_{11} & G_{12} \\ G_{21} & G_{22} \end{pmatrix} \begin{pmatrix} x_2(t) \\ u_2(t) \end{pmatrix} = \begin{pmatrix} (G_{11}x)(t) + (G_{12}u)(t) \\ (G_{21}x)(t) + (G_{22}u)(t) \end{pmatrix} \tag{3.13}$$

On setting $u_2(t) = 0$; then (3.14) will implies

$$(G z_2(t)) \equiv \begin{pmatrix} (G x_2)(t) \\ (G u_2)(t) \end{pmatrix} = \begin{pmatrix} y_{11}(t) \\ y_{21}(t) \end{pmatrix} \tag{3.14}$$

where the functions $y_{11}(t)$ and $y_{21}(t)$ must be determined in order to know $(G_{11}x_2)(t)$ and $(G_{21}x_2)(t)$. By virtue of the equivalence (3.10), we set $u_2(t) = 0$, in (3.9) as:

$$\langle z_1(t), z_2(t) \rangle = \int_0^{t_f} \left\{ \frac{1}{2} x_1(t) Q x_2(t) + x_1(t) H \dot{x}_2(t) + \mu \dot{x}_1(t) \dot{x}_2(t) - \mu A x_1(t) \dot{x}_2(t) - \mu B u_1(t) \dot{x}_2(t) - \mu A x_1(t) \dot{x}_2(t) + \mu A^2 x_1(t) x_2(t) - \mu B u_1(t) \dot{x}_2(t) \right\} dt \tag{3.15}$$

On simplifying (3.15) further, we obtain

$$\langle z_1(t), z_2(t) \rangle = \int_0^{t_f} \{ x_1(t) [(\frac{1}{2} Q + \mu A^2) x_2(t) + (H - 2\mu A) \dot{x}_2(t)] + \dot{x}_1(t) [\mu \dot{x}_2(t)] \} dt \tag{3.16}$$

$$\langle z_1(t), z_2(t) \rangle = \int_0^{t_f} \{ x_1(t) y_{11}(t) + \dot{x}_1(t) y_{11}(t) + u_1(t) y_{21}(t) \} dt \tag{3.17}$$

The quantities $y_{11}(t)$ and $y_{21}(t)$ which satisfy (3.17) have to be determined. Towards this, we set

$$\alpha(t) = (\frac{1}{2} Q + \mu A^2) x_2(t) + (H - 2\mu A) \dot{x}_2(t) \tag{3.18}$$

$$\beta(t) = \mu \dot{x}_2(t) \tag{3.19}$$

where $\beta(t)$ is the derivative of $\alpha(t)$ from (3.16) and $\alpha(t) = y_{11}(t)$ while $\beta(t) = \dot{y}_{11}(t)$ then, $\alpha(t) - y_{11}(t)$ and $\beta(t) - \dot{y}_{11}(t)$ are continuous functions on $[0, t_f]$ and for $x_1(\cdot) \in G_1[0, t_f]$ such that $x_1(0) = 0 = x_1(t_f)$, we have

$$\int_0^{t_f} \{ x_1(t) [\alpha(t) - y_{11}(t)] + \dot{x}_1(t) [\beta(t) - \dot{y}_{11}(t)] \} dt = 0, \tag{3.20}$$

so that by Lemma 1, $\beta(t) - \dot{y}_{11}(t)$ is differentiable on $[0, t_f]$ with

$$\frac{d}{dt} [\beta(t) - \dot{y}_{11}(t)] = \alpha(t) - y_{11}(t). \tag{3.21}$$

This implies that

$$\dot{y}_{11}(t) - y_{11}(t) = \dot{\beta}(t) - \alpha(t) \tag{3.22}$$

Using (3.18) and (3.19) in (3.22), we obtain

$$\dot{y}_{11}(t) - y_{11}(t) = \mu \ddot{x}_2(t) - [(\frac{1}{2} Q + \mu A^2) x_2(t) + (H - 2\mu A) \dot{x}_2(t)] \tag{3.23}$$

Now from (3.17), we obtain

$$u_1(t) y_{12}(t) = u_1(t) [-2\mu B \dot{x}_2(t)] \tag{3.24}$$

From which we obtain by virtue of (3.14)

$$y_{21}(t) = (G_{21}x_2)(t) = -2\mu B \dot{x}_2(t); \quad 0 \leq t \leq t_f \tag{3.25}$$

To determine $y_{11}(t)$, the second order differential equation (3.23) must be solved. For this purpose, let

$$f(t) = \mu \ddot{x}_2(t) - [(\frac{1}{2} Q + \mu A^2) x_2(t) + (H - 2\mu A) \dot{x}_2(t)] \tag{3.26}$$

Then, (3.23) takes the form

$$\dot{y}_{11}(t) - y_{11}(t) = f(t); \quad 0 \leq t \leq t_f \tag{3.27}$$

and in order to obtain a unique solution of (3.27), we append

the initial conditions

$$y_{11}(0) = n_0 \text{ and } \dot{y}_{11}(0) = m_0 \quad (3.28)$$

where the quantities n_0 and m_0 need to be determined.

In view of this, let $F(s) = \mathcal{L}\{f(t)\}$ and $y_{11}(s) = \mathcal{L}\{y_{11}(t)\}$ denote the Laplace transforms of $f(t)$ and $y_{11}(t)$ respectively, then by taking the Laplace transform of (3.27) and (3.28), we get

$$s^2 y_{11}(s) - y_{11}(s) - n_0 s - m_0 = F(s) \quad (3.29)$$

We proceed to determine the quantities $y_{11}(t), n_0$ and m_0 appearing in (3.29) solving (3.29) and the resulting constants therein eliminated, we obtain the following tidier form:

$$\begin{aligned} y_{11}(t) = & -\mu \dot{x}_2(0) \text{Sinh}(t_f) + \int_0^{t_f} \mu \dot{x}_2(s) \text{Cosh}(t_f - s) ds - \\ & \int_0^{t_f} \text{Sinh}(t_f - s) \times \left[\left(\frac{1}{2} Q + \mu A^2 \right) x(s) + (H - 2\mu A) \dot{x}_2(s) \right] ds \\ & + \left[\left(\frac{1}{2} Q + \mu A^2 \right) x_2(0) + (H - 2\mu A) \dot{x}_2(0) \right] \times \text{Cosht} \\ & + \frac{\text{Sinh } t}{\text{Sinh } t_f} \left\{ \left(\frac{1}{2} Q + \mu A^2 \right) x_2(t_f) + (H - 2\mu A) \dot{x}_2(t_f) \right\} \\ & - \left[\left(\frac{1}{2} Q + \mu A^2 \right) x_2(0) + (H - 2\mu A) \dot{x}_2(0) \right] \text{Cosh } t_f + \mu \dot{x}_2(0) \text{Sinh } t_f \\ & - \int_0^{t_f} \mu \dot{x}_2(s) \text{Cosh}(t_f - s) ds \\ & + \int_0^{t_f} \text{Sinh}(t_f - s) \left[\left(\frac{1}{2} Q + \mu A^2 \right) x(s) + (H - 2\mu A) \dot{x}_2(s) \right] ds; \\ & 0 \geq t \geq t_f \end{aligned} \quad (3.30)$$

Thus, the first column of the operator, G_{11} and G_{21} are uniquely determined by virtue of (3.30) and (3.25) respectively.

Repeating the same arguments, we obtain the second column of the operator, G_{12} and G_{22} , by setting $x_2(t) = 0$ in (3.9) from which we obtain

$$\begin{aligned} y_{12}(t) = & 2\mu ABu_2(0) \text{Cosht} - \int_0^{t_f} 2\mu ABu_2(s) \text{Sinh}(t_f - s) ds \\ & + \frac{\text{Sinh } t}{\text{Sinh } t_f} \left\{ 2\mu ABu_2(t_f) - 2\mu ABu_2(0) \text{Cosh}(t_f) \right\} \\ & + \int_0^{t_f} 2\mu ABu_2(s) \text{Sinh}(t_f - s) ds; \quad 0 \geq t \geq t_f \end{aligned} \quad (3.31)$$

$$y_{22}(t) = \left(\frac{1}{2} R + \mu B^2 \right) u_2(t); \quad 0 \geq t \geq t_f \quad (3.32)$$

This concludes the proof of Theorem 3.1.

4. Implementation of the Constructed Operator in ECGM Algorithm

The following linear regulator problems were solved using the ECGM algorithm described above.

Problem (P1):

$$J(x, t_f, u(\cdot)) = \frac{1}{2} x^2(1) + \frac{1}{2} \int_0^1 \{x^2(t) + u^2(t)\} dt$$

subject to

$$\dot{x}(t) = 2x(t) + 3u(t), \quad 0 \leq t \leq t_f,$$

$$x_0 = 5 \text{ and } u = 3$$

Problem (P2):

$$J(x, t_f, u(\cdot)) = \frac{1}{2} x^2(1) + \frac{1}{2} \int_0^1 \{x^2(t) + u^2(t)\} dt$$

subject to

$$\dot{x}(t) = -3x(t) - 2u(t), \quad 0 \leq t \leq t_f,$$

$$x_0 = 5 \text{ and } u = 3$$

Problem (P3):

$$J(x, t_f, u(\cdot)) = \frac{1}{2} x^2(1) + \frac{1}{2} \int_0^1 \{x^2(t) + u^2(t)\} dt$$

subject to

$$\dot{x}(t) = 5x(t) + 3u(t), \quad 0 \leq t \leq t_f,$$

$$x_0 = 2 \text{ and } u = 3$$

5. Conclusion

It follows from here that, while [5] constructed an operator for CLRP, [2] focuses on same class of optimal control problem but with delay parameter in the state variable. The construction of this control operator, G, helps to bridge the gap between Bolza form of CLRP and Lagrange as well as Mayer Forms of CLRP. This makes the construction of the operator very important and relevant in that, it takes cares of the variations in CLRP with or without boundary penalty variable. In determining the convergence of the problems under review, the Gradient Norm was used and set at 10^{-10} and the Function Value was also considered at the point when the function value starts to repeat itself. From the Tables 1 and 2, Problems (P1) and (P2) converged at the 9th iteration with gradient norms 2.98910531e-9 and 1.1127029e-9 respectively while problem (P3) converged at the 23rd iteration with gradient norm 3.5220424e-9.

Table 1. Table of Result (P1): At Penalty Parameter $\mu = 0.01$

Iteration	x	u	Function Value	Gradient Norm
0	5	3	62.61	13.4391964
1	3.06855136	1.81686131	23.4757448	3.1517481512
2	3.52472081	0.43686661	25.7370688	0.359818697e-1
3	3.51684235	0.43415944	25.6197708	0.931618947e-2
4	3.51831431	0.41523148	25.6154572	0.973626732e-3
5	3.51810621	0.41519251	25.6124570	0.268623676e-4
6	3.51811448	0.41514042	25.6124559	0.260293953e-4
7	3.51810917	0.41513334	25.6123726	0.125576612e-6
8	3.51810923	0.41513332	25.6123724	0.865124831e-7
9	3.51810923	0.41513317	25.6123721	0.298910531e-8

Table 2. Table of Result (P2): At Penalty Parameter $\mu = 0.1$

Iteration	x	u	Function Value	Gradient Norm
0	5	3	103.1	46.8017094
1	2.98780488	1.99593496	21.7972618	0.99496052
2	2.14743813	1.59661256	20.4926565	0.15828147
3	2.14179312	1.58006784	20.8593797	0.38950417e-2
4	2.14155486	1.58208179	20.8700589	0.20906703e-2
5	2.14118383	1.58222782	20.8657692	0.19568572e-4
6	2.14118198	1.58222303	20.8657082	0.12744208e-4
7	2.14118479	1.58221888	20.8657205	0.20471286e-6
8	2.14118482	1.58221893	20.8657211	0.44541463e-7
9	2.14118481	1.58221892	20.8657211	0.11127029e-8

Table 3. Table of Result (P3): At Penalty Parameter $\mu = 0.1$

Iteration	x	u	Function Value	Gradient Norm
0	2	3	53.1	25.8069758
1	-0.74256891	1.04102221	2.22132825	1.35523175
2	-1.53027208	2.02974998	9.04736912	1.55635029
3	-1.51972594	2.27465778	9.85321115	0.26905925
4	-1.49421917	2.28021637	9.70451487	0.12727048
5	-1.46725095	2.24614227	9.38654563	0.11204496
6	-1.46919089	2.23082279	9.33631847	0.1942723e-1
7	-1.47122794	2.23059326	9.34870783	0.10412658e-1
8	-1.47337643	2.23362683	9.37512075	0.83226081e-2
9	-1.47315285	2.23465202	9.37782731	0.14227137e-2

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